



Center for Spatially Integrated Social Science

Variogram Analysis

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Outline

- Geostatistical Perspective
- Variogram and Correlogram
- Modeling Variograms

Geostatistical Perspective

Geostatistical Perspective

➤ Continuous Spatial Index

- $\{ Z(s): s \in D \}$

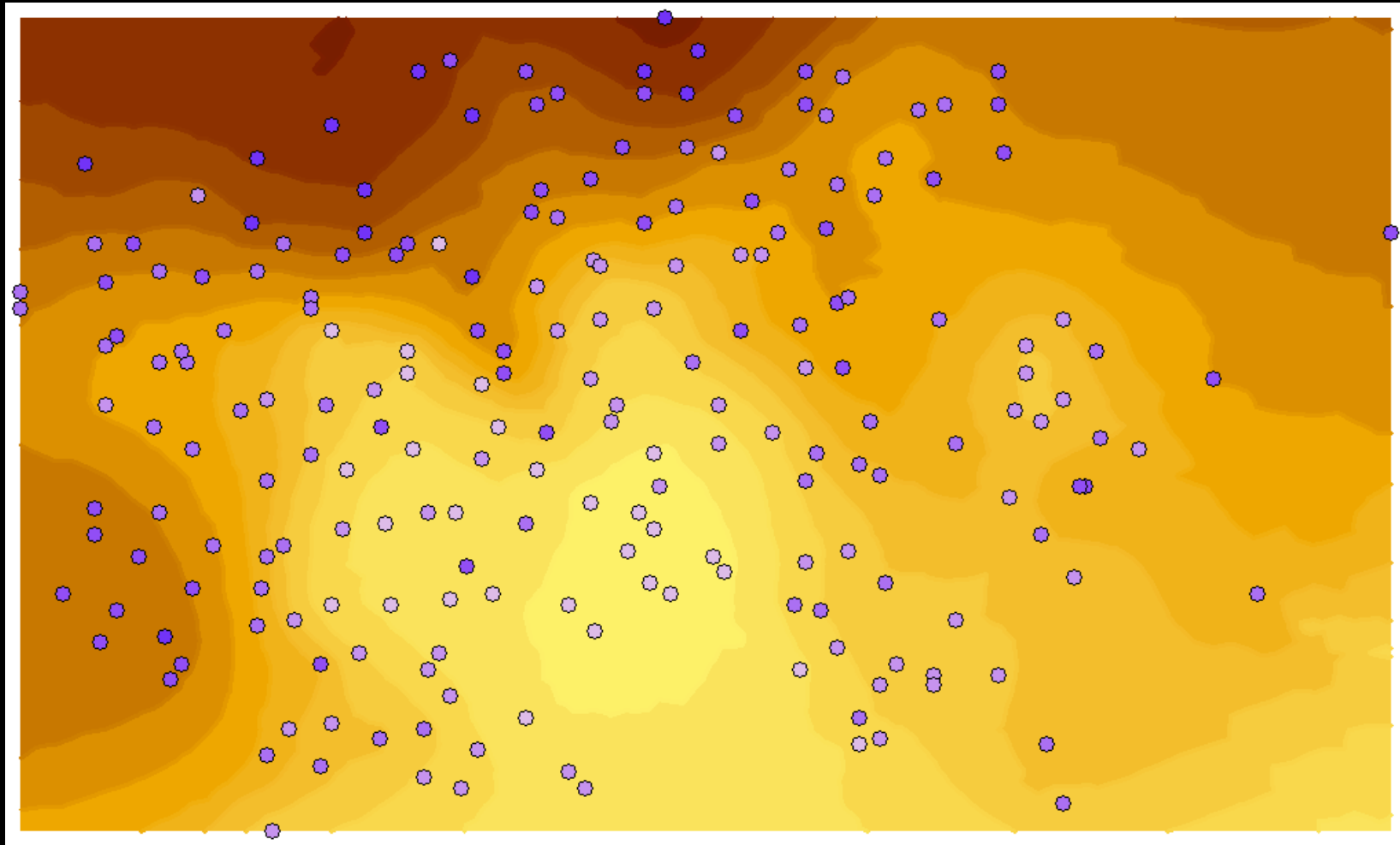
- s is spatial index
- continuous in R^d

- sample of spatial locations

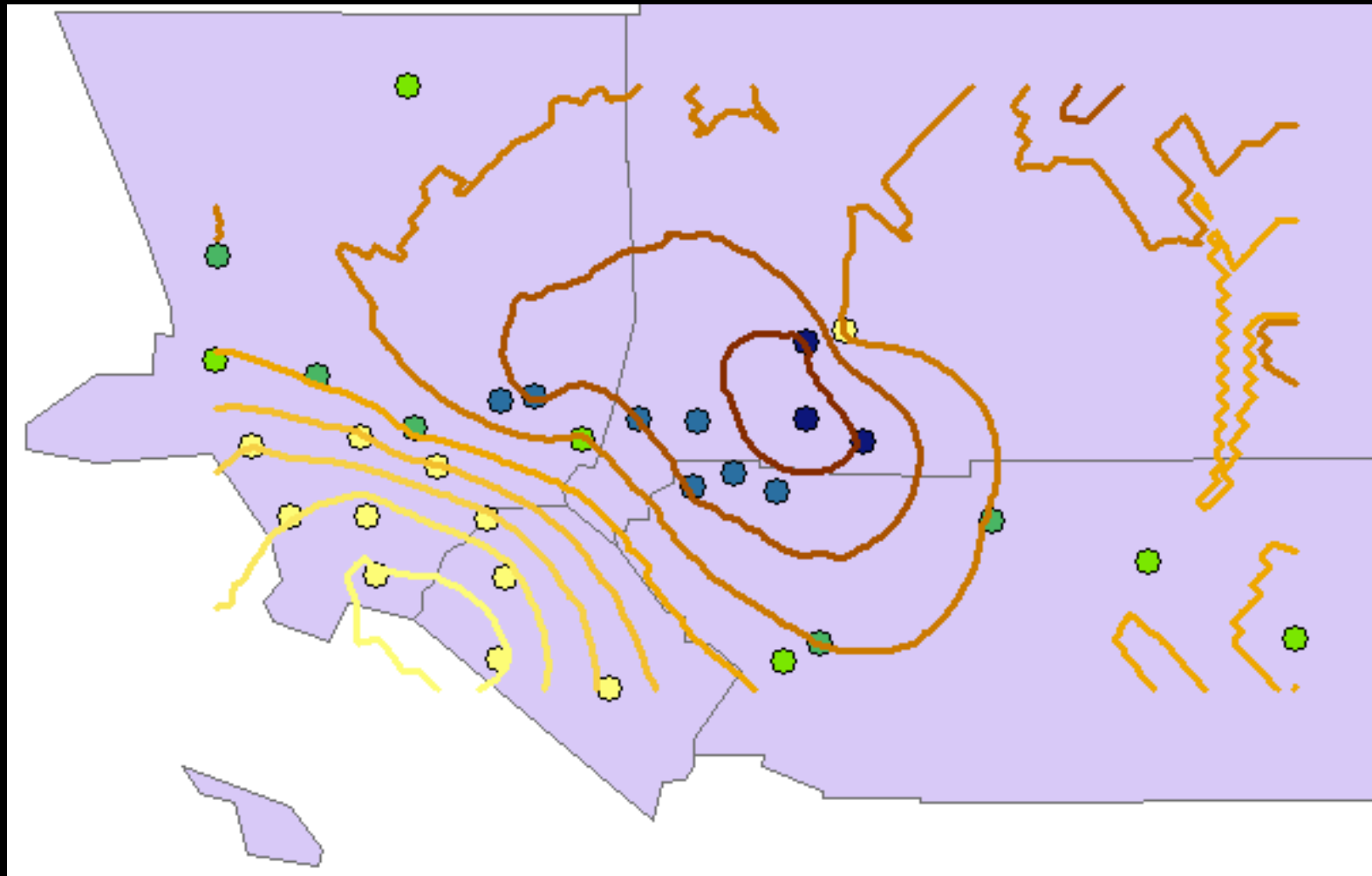
- $\{ s_1, s_2, \dots, s_N \} =$ locations
- $\{ Z(s_1), Z(s_2), \dots, Z(s_N) \} =$ random variable at locations

➤ Spatial Random Field

- model for a continuous spatial process
- field data model



Residential Sales Price, Baltimore MD (1980)
sample points (darker is higher) and contours



Air Quality (Ozone) in Los Angeles Basin
Contours from Spherical Variogram Interpolation

Conceptual Framework

➤ Equilibrium = Stationarity

- stochastic process
 - not multiple realizations, but **single realization**
 - the map is a single data point
- notion of stability needed to relate sample (= single observation) to population
 - even though there is only one data point, act **as if there are multiple observations**

Concept

3rd "realization"



2nd "realization"



reality considered
as a 1st "realization"
of the regionalized variable

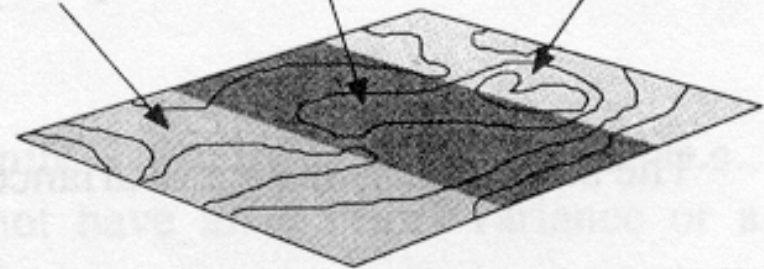


Application
(stationarity hypothesis)

2nd realization

3rd realization

1st realization



Implications of Spatial Stationarity (source Y. Pannatier, Variowin)

Moment Conditions

➤ Constraints on Variability

▪ ergodicity

- average over single realization same as over all possible
- whether you see one or many maps, information is same

▪ moments must exist

- no infinite variance

▪ moments must be “regular” over space

- restrictions on **heterogeneity**, on **range** of dependence

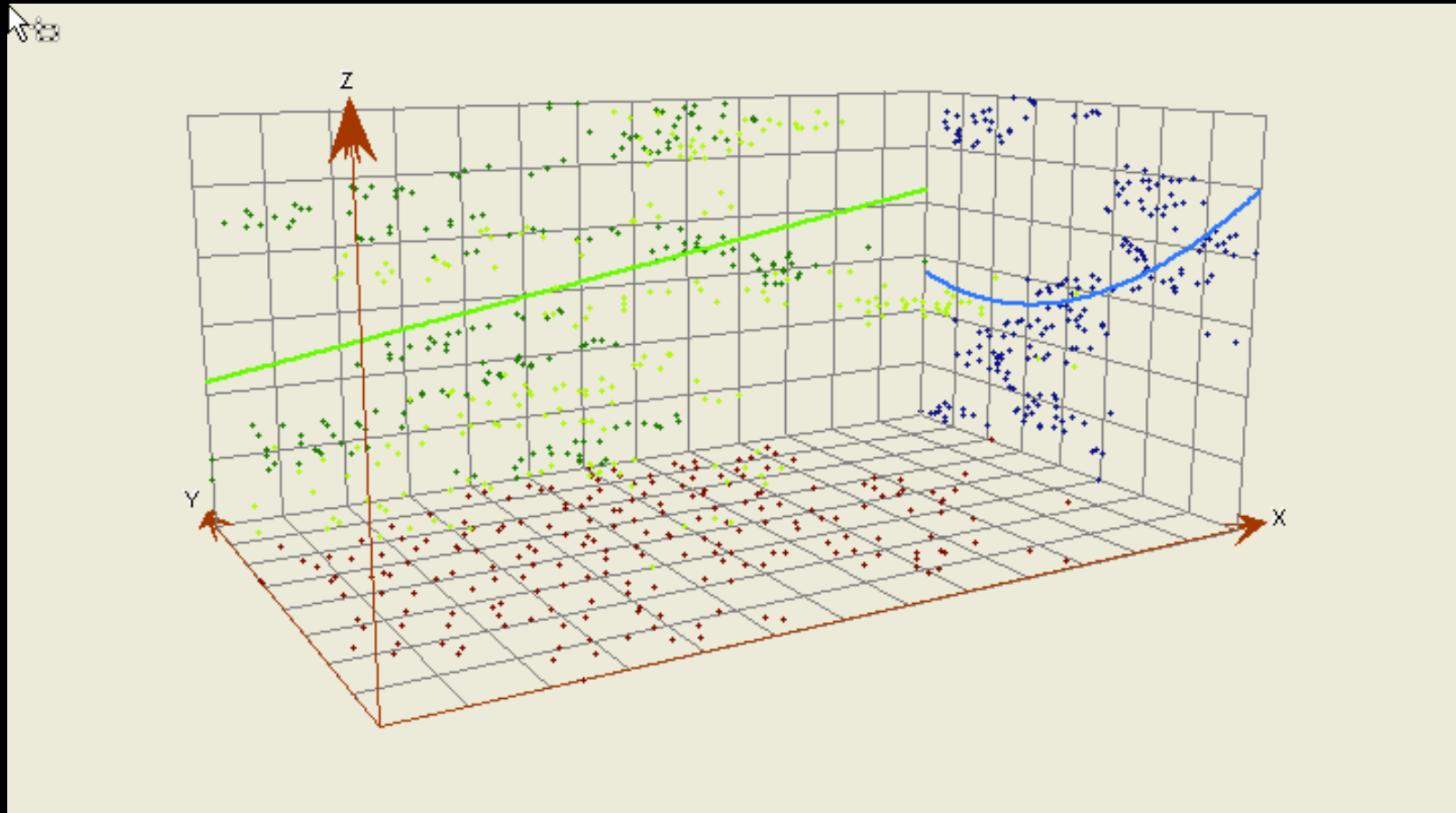
Spatial Stationarity

➤ Strict Stationarity

- invariance of joint probability density function under spatial shift ("translation")
 - $\{ z(x_1), \dots, z(x_k) \}$ and $\{ z(x_{1+h}), \dots, z(x_{k+h}) \}$
 - information about process the same no matter where it is obtained

➤ Moment Stationarity

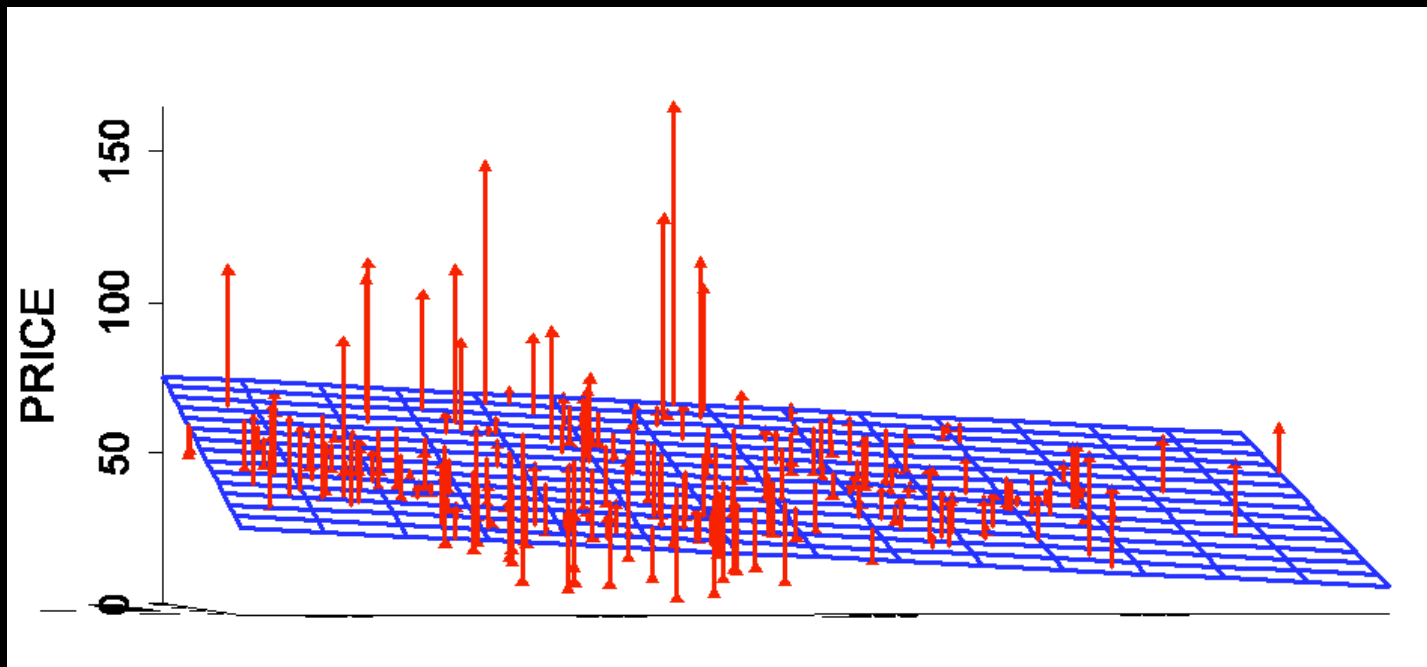
- moments invariant under shift
 - constant mean and constant variance
 - covariance only function of spatial separation h



Baltimore House Sales Prices
Spatial Trend Analysis

Linear Spatial Trend

Baltimore House Prices



$$P = -166.02 - 0.148 X + 0.634 Y \quad R^2 = 0.27$$

Variogram and Correlogram

Variogram

➤ Intrinsic Hypothesis

- **no spatial trend**
 - if there is a trend, take it out
 - **residuals** have no trend by construction (mean = 0)
- **variance constant**
- **variance of first differences only a function of displacement**
 - **Var** { $Z[s+h] - Z[s]$ }
 - how does the **variability of the difference** change with **h**

Semi-Variogram

➤ General Variogram Function

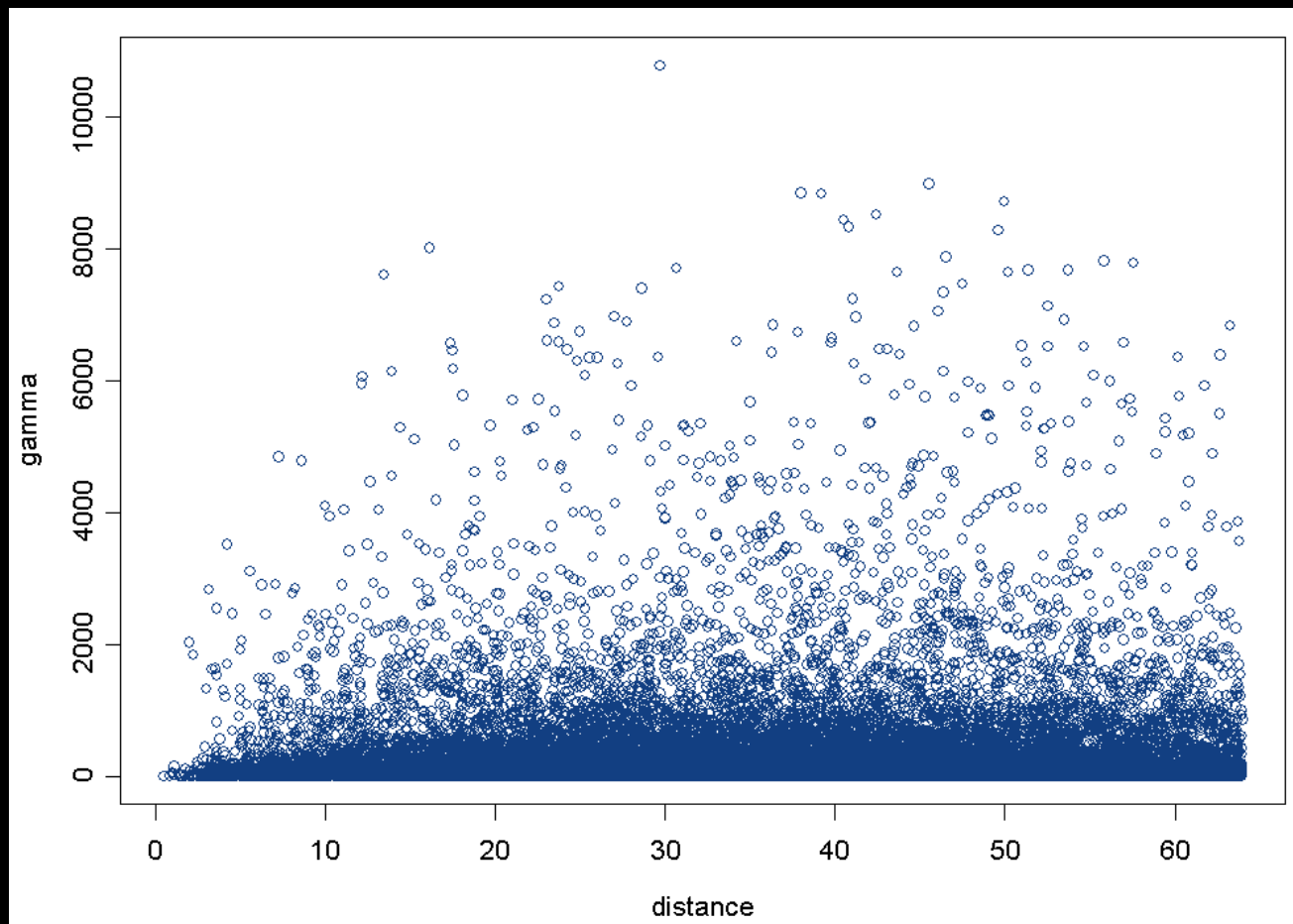
- $2\gamma(h) = \text{Var} [Z(s+h) - Z(s)]$
 - note the factor 2, hence $\gamma(h)$ is 1/2 of variogram or semivariogram

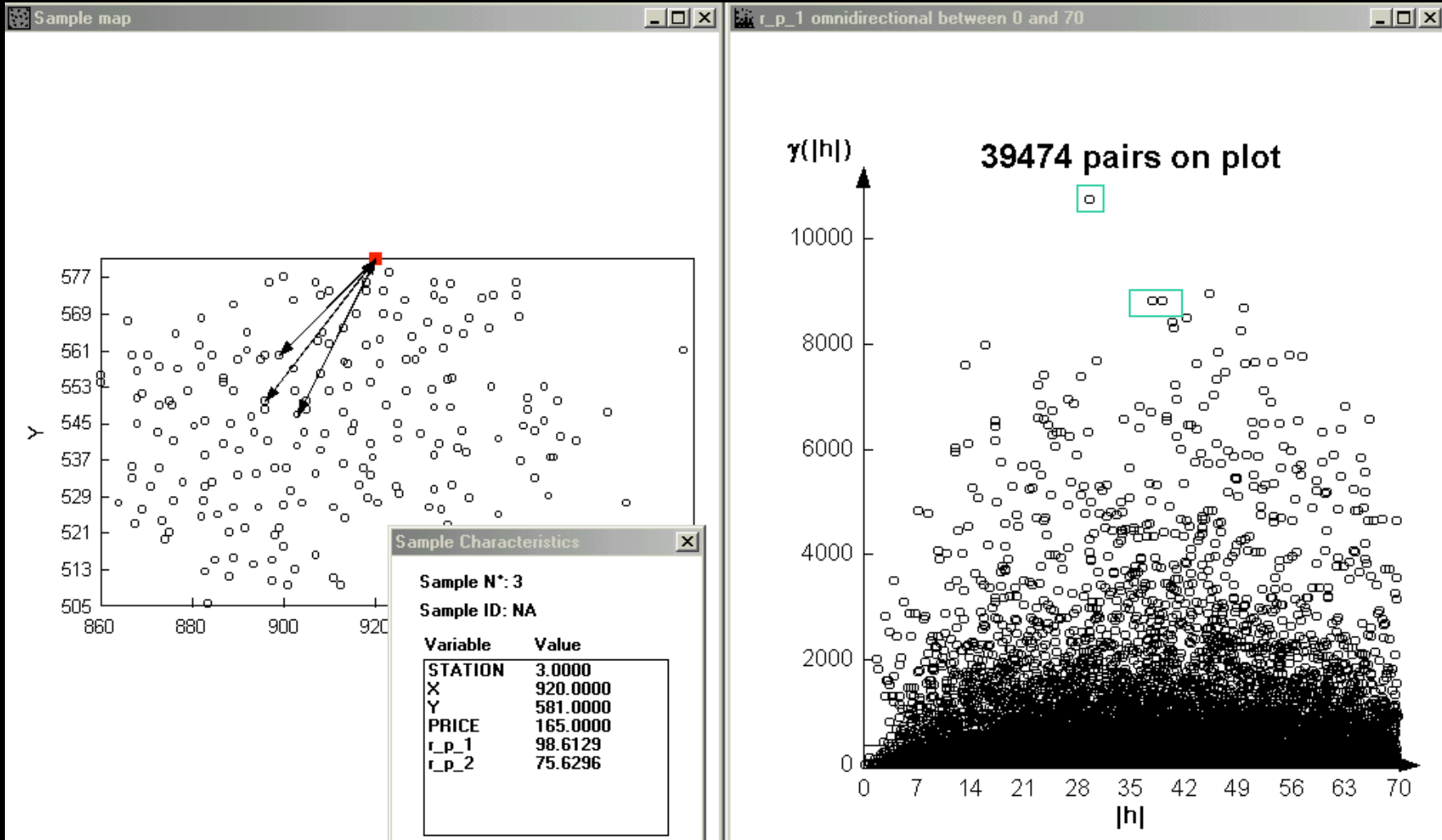
➤ Regular Case

- for intrinsic hypothesis, $E[Z(s+h)] = E[Z(s)]$
 - constant mean, hence $E[Z(s+h) - Z(s)] = 0$
 - $\text{Var}[Z(s+h) - Z(s)] = E\{[Z(s+h) - Z(s)] - E[Z(s+h) - Z(s)]\}^2$
- $2\gamma(h) = E [Z(s+h) - Z(s)]^2$
 - average of squared differences

(Semi)Variogram Cloud Plot

(residuals of Baltimore trend surface)





Brushing a Variogram Cloud Plot
identification of potential outliers

Estimating a Variogram

➤ Method of Moments

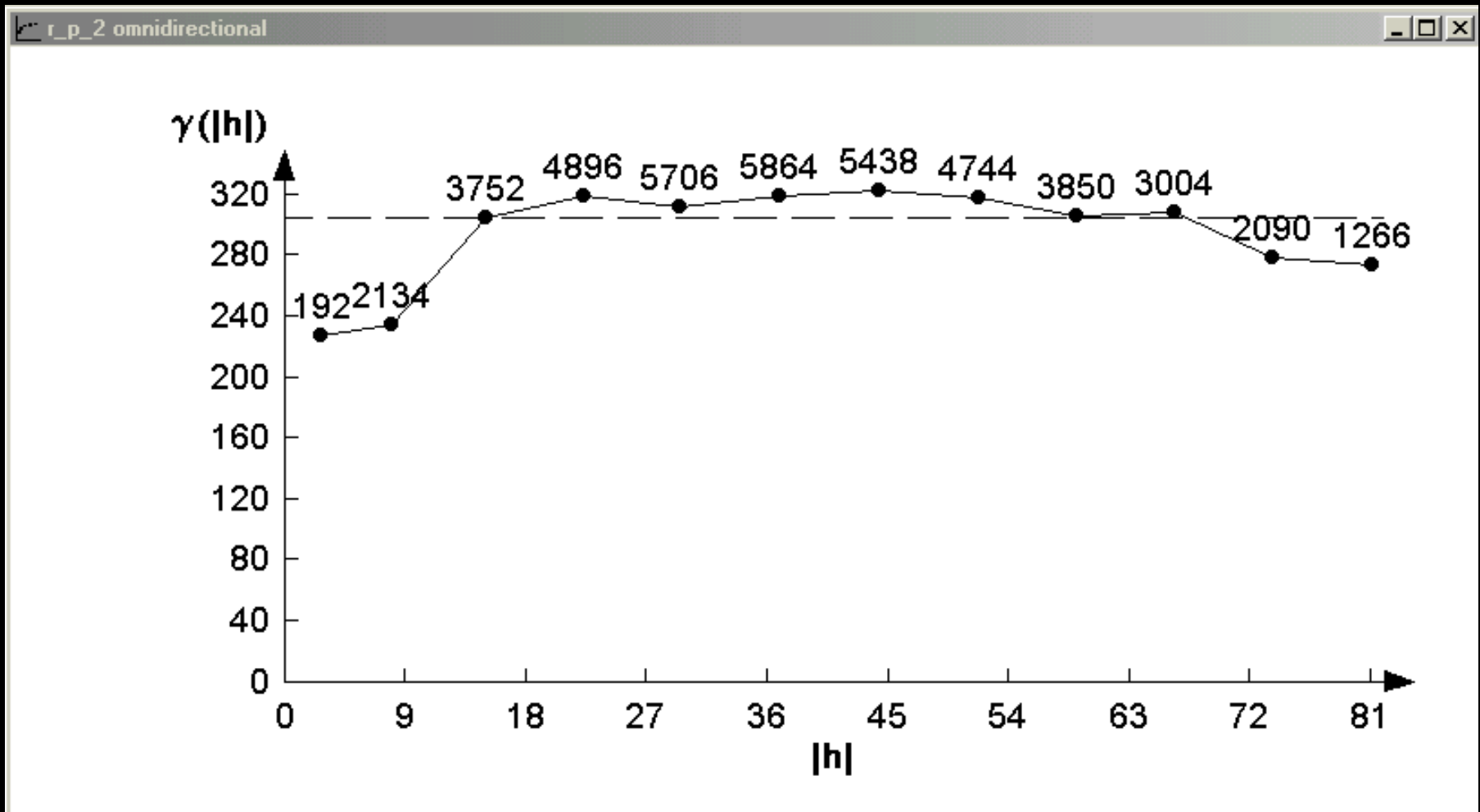
➤ $2\hat{\gamma}(h) = \{ 1 / | N(h) | \} \times \sum_{N(h)} [Z(s+h)-Z(s)]^2$

- average of squared differences by distance bin

- h : distance bin
- $N(h)$: number of pairs in distance bin for h

➤ Rules of Thumb

- at least 30 pairs in each bin
- $h < D / 2$ (D is max distance)
 - distance of reliability



Variogram for residuals from second order trend surface

Covariogram

- Second Order Stationarity
 - covariance “regular” over space
 - assumptions stronger than for (semi) variogram
 - define regularity for second order moments
- Covariogram
 - $C(h) = \text{Cov} [Z(s+h), Z(s)]$
 - covariance as a function of “distance” h
 - $C(0) = \text{Var} [Z(s)]$
 - at zero distance, covariance = variance of process

Correlogram

➤ Correlogram

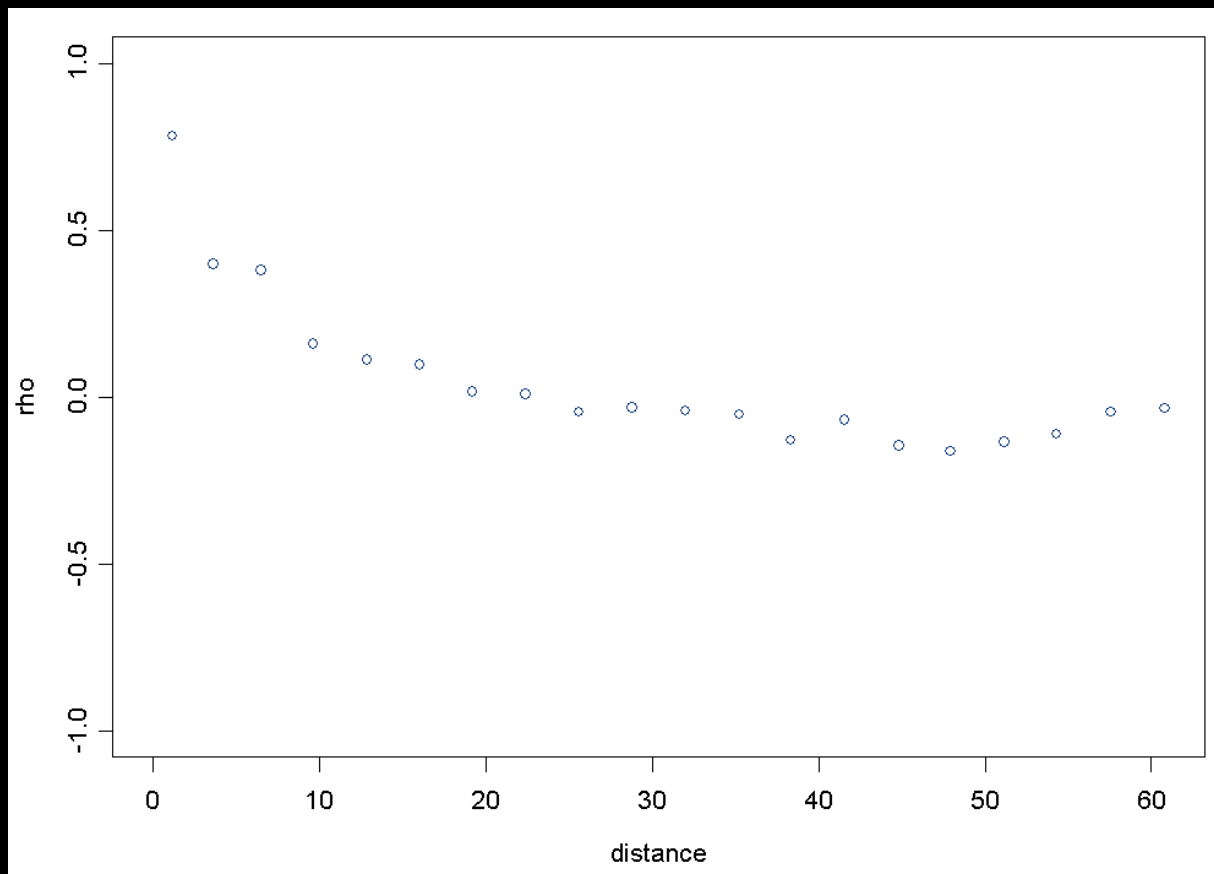
- $\rho(h) = C(h) / C(0)$
- notion of an “autocorrelation” function
- covariogram $C(h)$ standardized by variance $C(0)$

➤ Semi-Variogram and Correlogram

- $2\gamma(h) = E \{ [Z(s+h) - Z(s)]^2 \}$
 $= 2 E \{ Z(s) \}^2 - 2 E \{ Z(s+h) \cdot Z(s) \}$
- $\gamma(h) = C(0) - C(h)$
- correlogram decreases with distance
- semivariogram increases with distance

Correlogram

(Baltimore trend surface residuals)



Modeling Variograms

Range and Sill

➤ Limit Behavior

- $\gamma(h) = C(0) - C(h)$
- as $h \rightarrow \infty$, $C(h) \rightarrow 0$ or $\gamma(h) \rightarrow C(0)$
 - limit on range of dependence

➤ Sill

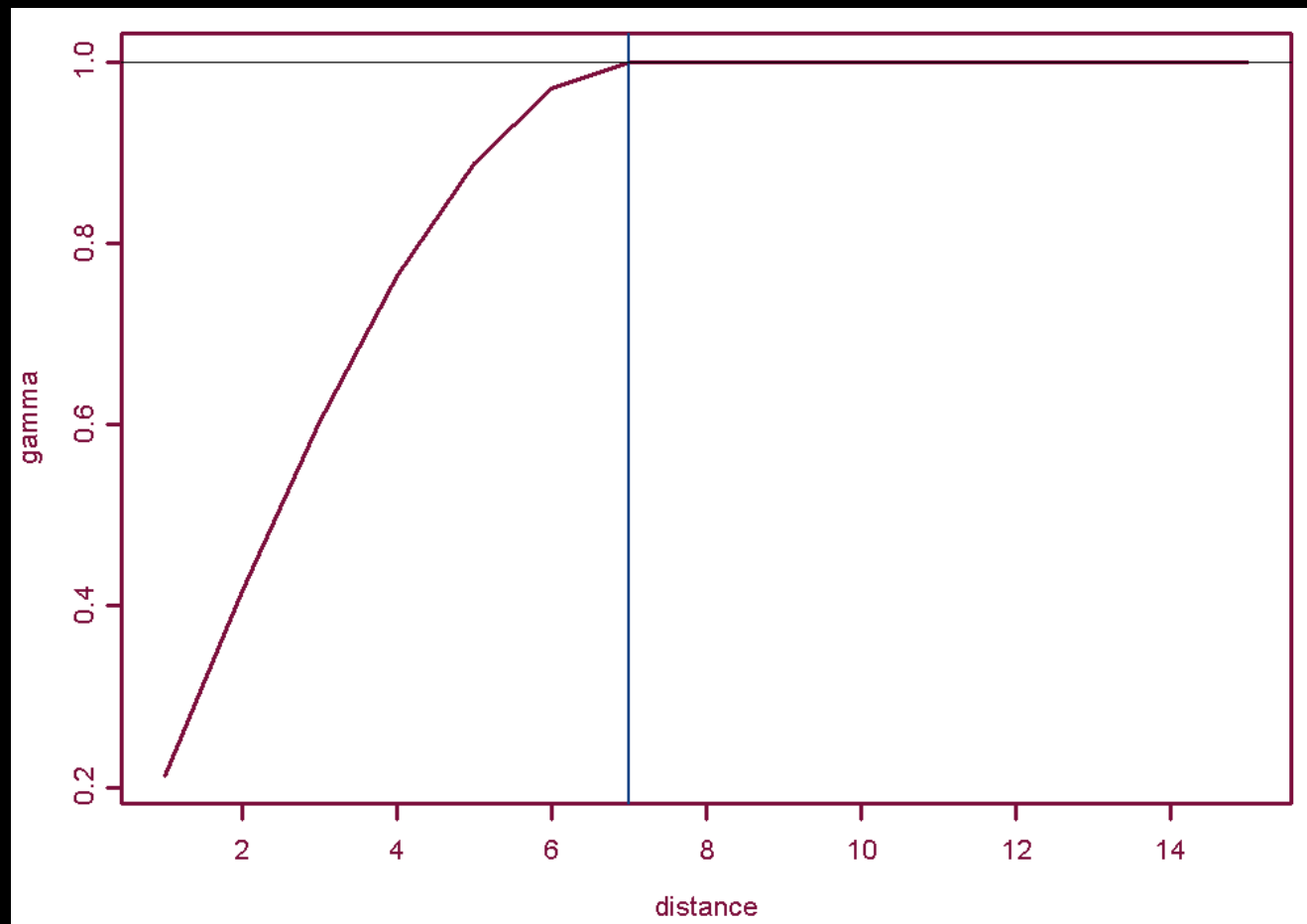
- $C(0)$ is sill = $\text{Var}[Z(s)]$

➤ Range

- h s.t. $\gamma(h) = C(0)$ is range
 - use range in **spatial sampling**
 - range is distance beyond which there is no spatial autocorrelation

Theoretical Spherical Variogram

sill = 1



range = 7

Nugget Effect

- Behavior Near $h = 0$
 - $\gamma(h) \rightarrow c_0 > 0$ as $h \rightarrow 0$
 - not possible mathematically
 - $\gamma(0) = 0$ by definition
- Interpretation of Nugget Effect
 - measurement error for $h < h_{\min}$
 - scale problems

Valid Variogram Models

- Moment Conditions
 - $C(\bullet)$ positive definite $\forall h$
 - $\gamma(\bullet)$ negative definite $\forall h$
- Isotropic Variogram Models
 - $\gamma(h, \theta) = 0$ for $h = 0$
 - $\gamma(h, \theta)$ same in all directions
- Non-Isotropic Models
 - directional
 - different variogram for different directions

Examples of Variogram Models

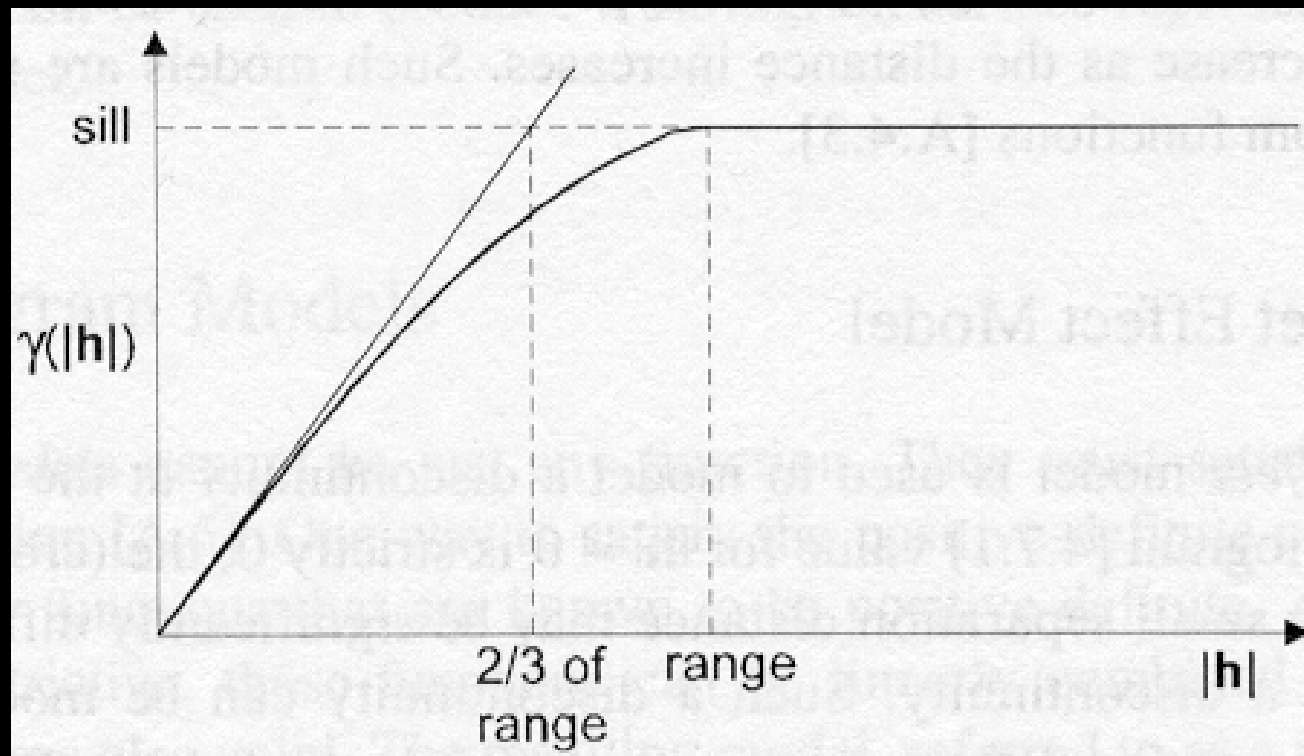
➤ Spherical

- $\gamma(h, \theta) = c_0 + c_s \{1.5h/a - 0.5(h/a)^3\}$
for $0 < h \leq a$
 $= c_0 + c_s$ for $h \geq a$
- c_0 = nugget effect, $c_0 + c_s$ = sill, a = range

➤ Exponential

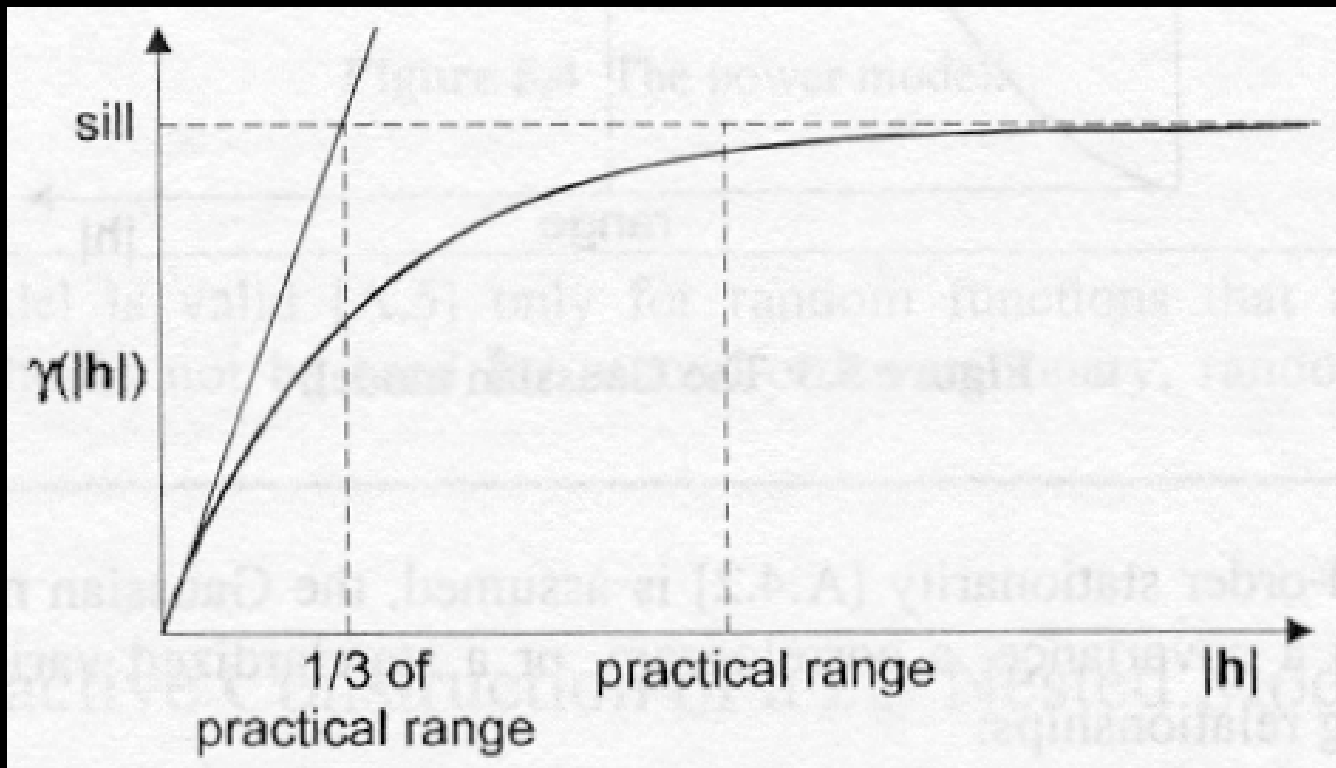
- $\gamma(h, \theta) = c_0 + c_s \{1 - e^{-(3h/a)}\}$
- a is "practical range" 95% of asymptotic range

Spherical Variogram



Source Pannatier 96

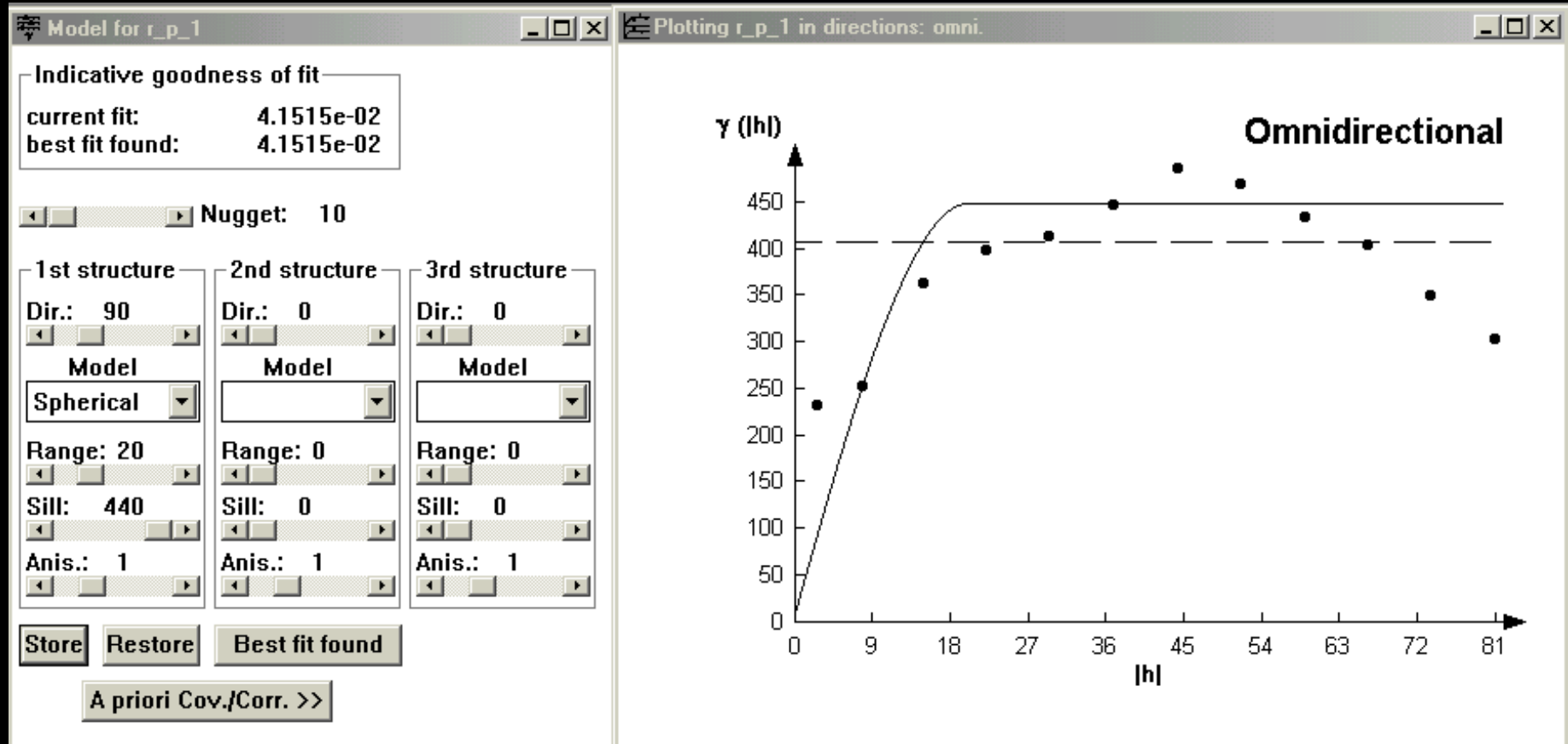
Exponential Variogram



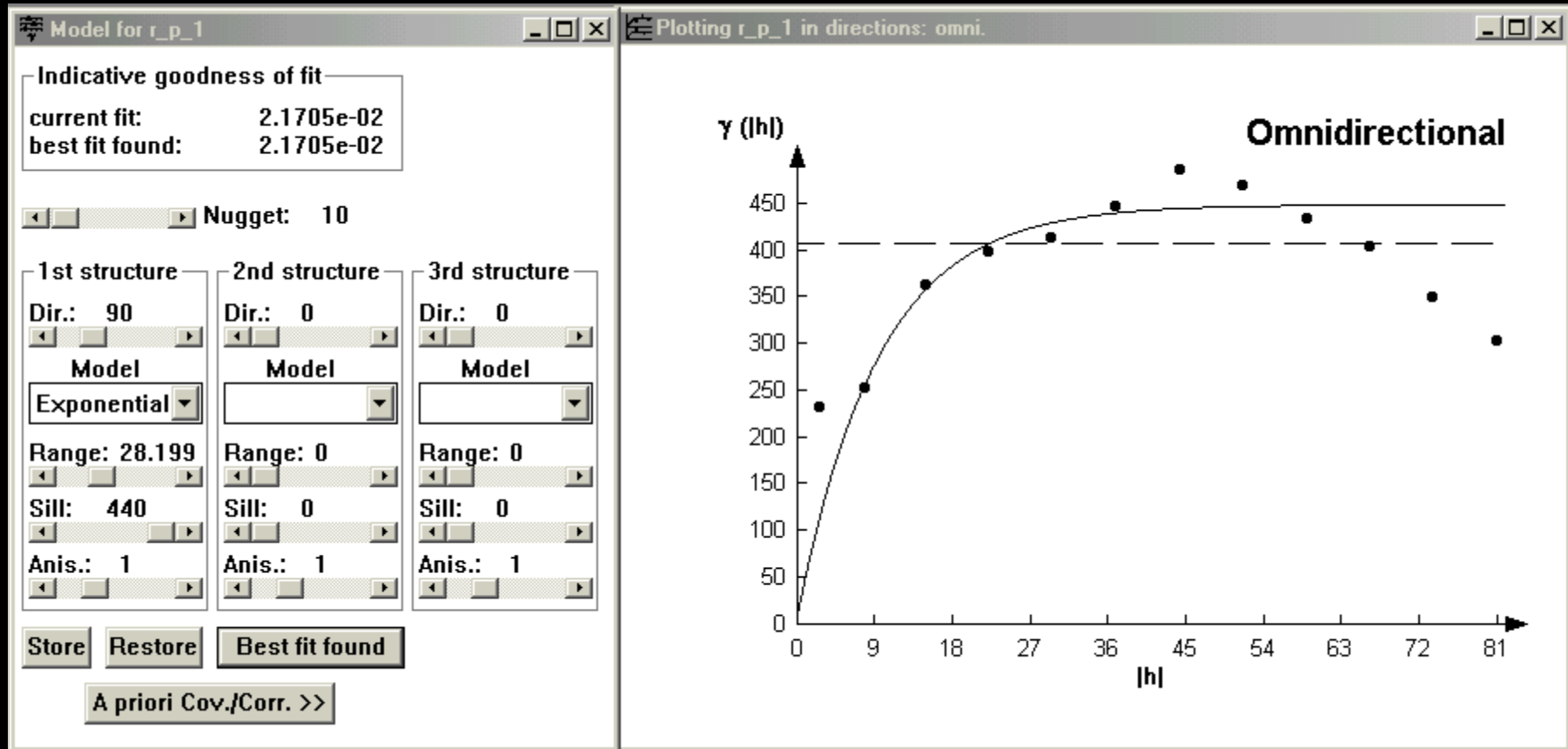
Source Pannatier 96

Fitting a Variogram Model

- Parameter Estimates
 - from empirical variogram
 - methods
 - nonlinear least squares
 - weighted least squares
 - “eyeball”
- Parameters must satisfy constraints
 - valid parameter space



Fit of Spherical Variogram Model to Baltimore Residuals



Fit of Exponential Variogram Model to Baltimore Residuals

